

COVID in MA analyzer

Introduction and citations

Data sources

I'm using the wiki compilation at https://en.wikipedia.org/wiki/2020_coronavirus_pandemic_in_Massachusetts which gives confirmed cases of COVID-19 in Massachusetts, according ultimately to the daily updates from the MA Dept of Public Health, such as <https://www.mass.gov/doc/covid-19-cases-in-massachusetts-as-of-march-28-2020/download>.

Subsidiary factoids:

Population of Boston:

```
pop_yr=[2010, 2017];
pop_pp=[620702,685094];
pop_2020_bos=diff(pop_pp)/diff(pop_yr)*3+pop_pp(2)
```

```
pop_2020_bos = 7.1269e+05
```

Population of Massachusetts:

```
pop_yr=[2005, 2018];
pop_pp=[6.454e6,6.902e6];
pop_2020_ma=diff(pop_pp)/diff(pop_yr)*2+pop_pp(2)
```

```
pop_2020_ma = 6.9709e+06
```

Intent

Just to have an idea of when we should see a peak.

Data sets and structure

Data are given in the table Cov_Ma.

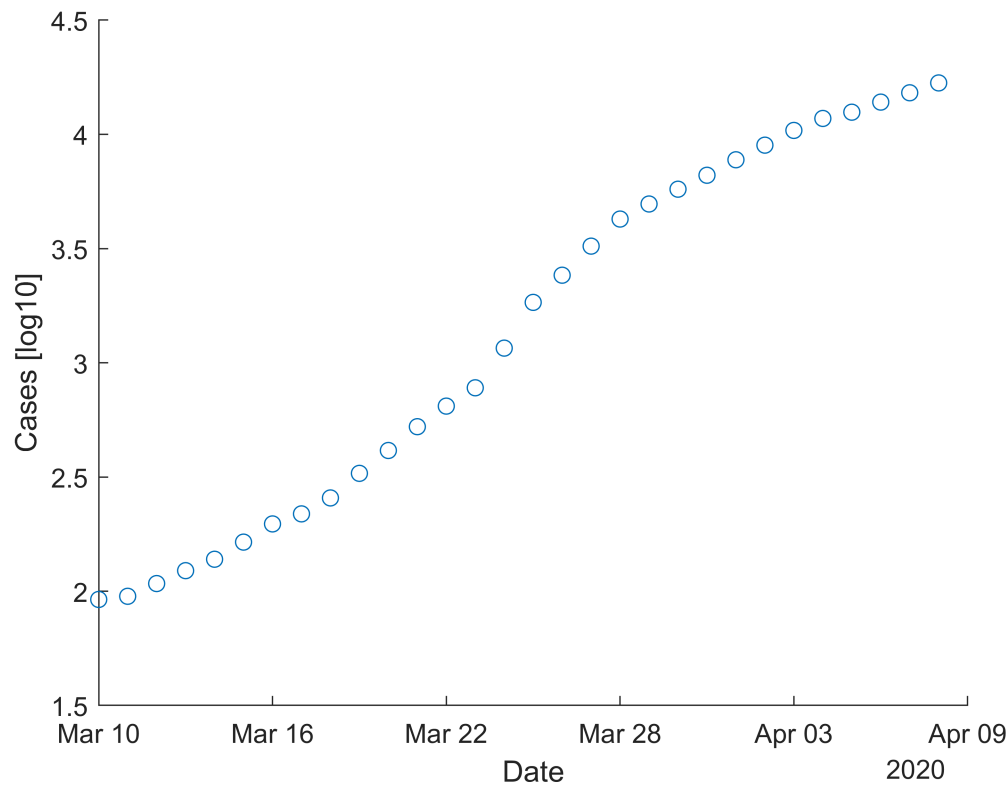
```
Cov_Ma.Properties.VariableNames
```

```
ans = 1x12 cell array
```

```
    {'Days_since_Mar_10'}    {'Cases'}    {'Date'}    {'Cases_change'}    {'Cases_growth'}    {'Tests'}    {'Tests_...
```

Just the facts, ma'am

```
figure
scatter(Cov_Ma.Date, log10(Cov_Ma.Cases))
ylabel('Cases [log10]')
xlabel('Date')
```



Fitting the data to a logistic curve

The logistic curve is given by

$$m_i = \frac{k_i}{1 + e^{-\gamma(t-t_0)}}$$

where m_i is a measure of COVID (I use identified cases and identified deaths), t is the number of days since March 10 (a reasonable measure of when community spreading started), t_0 is the midpoint of the growth (measured in days since March 10), and γ is a constant related to growth rate.

One must initialize the vector of constants with initial conditions to begin the minimization that finds the best-fitting array of constants. These are determined by trial and error. Good practice is to use a variety of IC to make sure the solution is robust. The ICs are in array `beta0_i`. MATLAB makes fitting the model pretty easy.

Flts

Cases

```
modelfun_c = @(b,x)b(1)./(1+exp(-(x-b(2))*b(3)))
```

`modelfun_c` = function_handle with value:

```
@(b,x)b(1)./(1+exp(-(x-b(2))*b(3)))
```

```
beta0_c = [18000 23 0.27];
```

```
mdl_c = fitnlm(Cov_Ma.Days_since_Mar_10,Cov_Ma.Cases,modelfun_c,beta0_c)
```

```
mdl_c =
```

```
Nonlinear regression model:  
y ~ F(b,x)
```

```
Estimated Coefficients:
```

	Estimate	SE	tStat	pValue
b1	21174	815.58	25.962	1.2478e-20
b2	24.175	0.36406	66.402	1.8716e-31
b3	0.24907	0.0089142	27.941	1.8396e-21

```
Number of observations: 30, Error degrees of freedom: 27  
Root Mean Squared Error: 252  
R-Squared: 0.998, Adjusted R-Squared 0.998  
F-statistic vs. zero model: 7.27e+03, p-value = 2.38e-39
```

Deaths

```
modelfun_d = @(b,x)b(1)./(1+exp(-(x-b(2))*b(3)))
```

```
modelfun_d = function_handle with value:
```

```
@(b,x)b(1)./(1+exp(-(x-b(2))*b(3)))
```

```
beta0_d = [10000 20 0.1];
```

```
mdl_d = fitnlm(Cov_Ma.Days_since_Mar_10,Cov_Ma.Deaths,modelfun_d,beta0_d)
```

```
mdl_d =
```

```
Nonlinear regression model:  
y ~ F(b,x)
```

```
Estimated Coefficients:
```

	Estimate	SE	tStat	pValue
b1	1253.8	511.42	2.4515	0.020981
b2	31.92	2.7399	11.65	4.8709e-12
b3	0.23707	0.021681	10.934	2.0197e-11

```
Number of observations: 30, Error degrees of freedom: 27  
Root Mean Squared Error: 11.9  
R-Squared: 0.99, Adjusted R-Squared 0.99  
F-statistic vs. zero model: 1.33e+03, p-value = 1.9e-29
```

Predictions

```
Pred_window=16;
```

```
Pred_days=0:max(Cov_Ma.Days_since_Mar_10)+Pred_window;
```

```
Pred_dates=(min(Cov_Ma.Date):days(1):max(Cov_Ma.Date)+days(Pred_window));
```

```
[Cov_Ma_Pred.cases, Cov_Ma_Pred.casesci]=predict(mdl_c,Pred_days','Prediction','observation');
```

```
[Cov_Ma_Pred.deaths, Cov_Ma_Pred.deathsci]=predict(mdl_d,Pred_days','Prediction','observation');
```

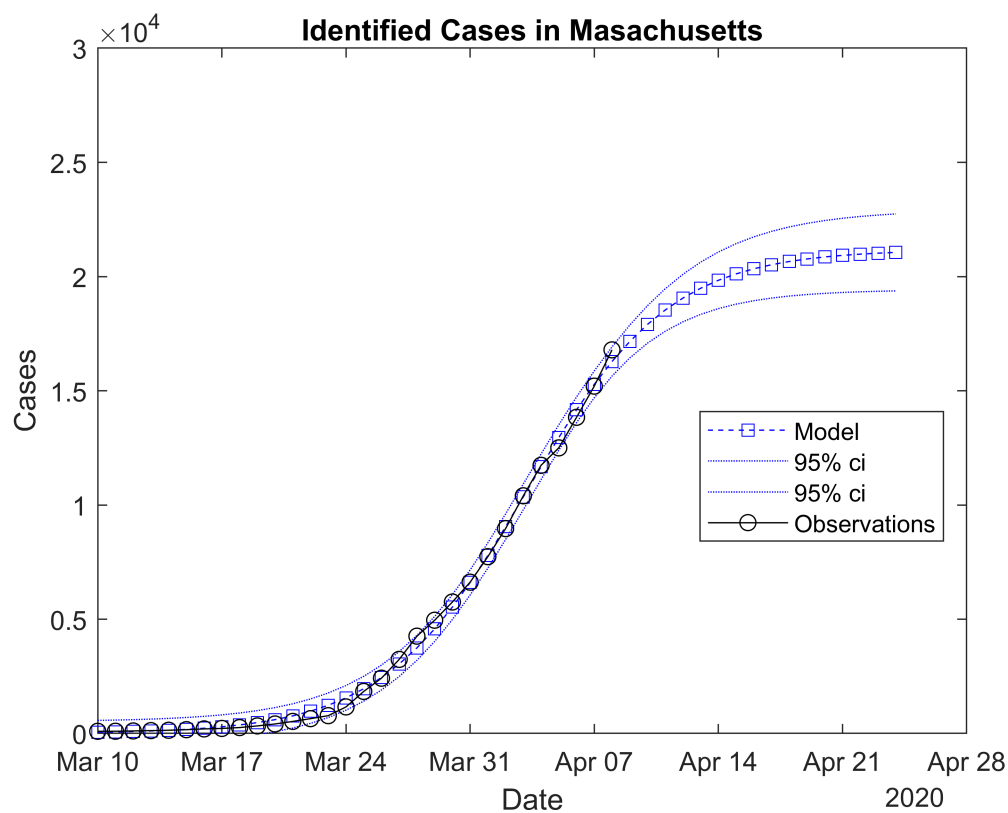
Graph the data and the models up

Cases

```
figure
plot(Pred_dates,Cov_Ma_Pred.cases,'--bs')
hold
```

Current plot held

```
plot(Pred_dates,Cov_Ma_Pred.casesci(:,1),':b')
plot(Pred_dates,Cov_Ma_Pred.casesci(:,2),':b')
plot(Cov_Ma.Date,Cov_Ma.Cases, '-ko')
ylabel('Cases')
xlabel('Date')
ylim([0,30000])
title('Identified Cases in Massachusetts')
legend('Model','95% ci','95% ci','Observations', 'location', 'best')
```



Deaths

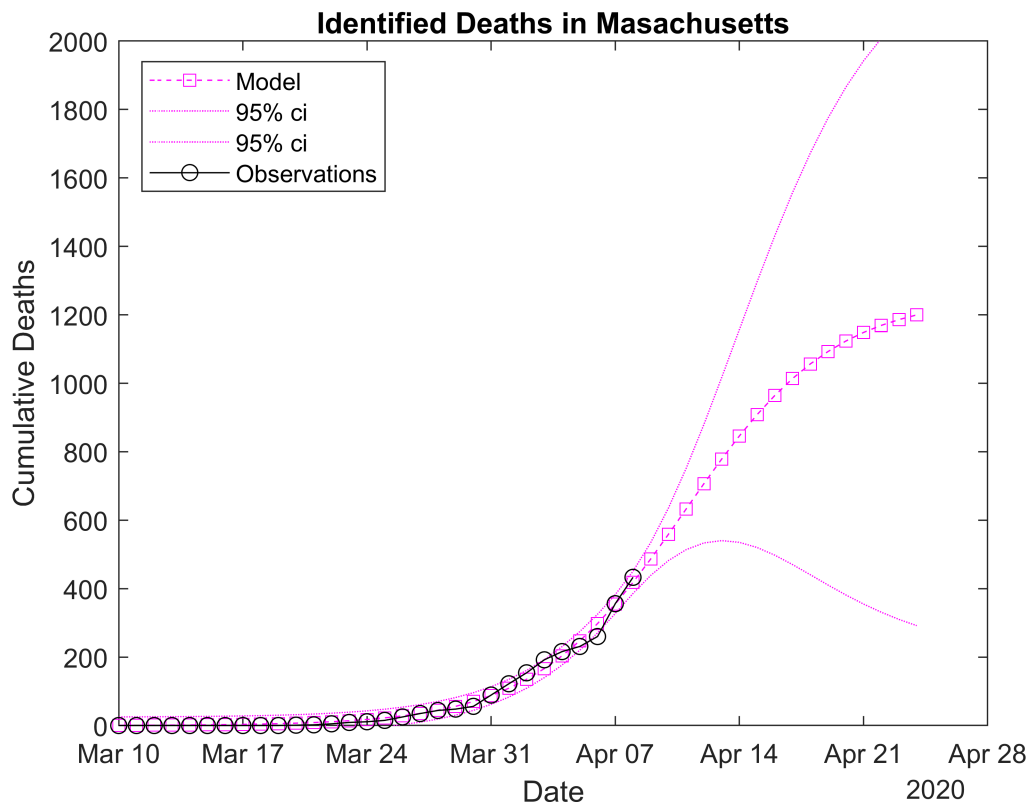
```
figure
plot(Pred_dates,Cov_Ma_Pred.deaths,'--ms')
hold
```

Current plot held

```

plot(Pred_dates,Cov_Ma_Pred.deathsci(:,1),'m')
plot(Pred_dates,Cov_Ma_Pred.deathsci(:,2),'m')
plot(Cov_Ma.Date,Cov_Ma.Deaths,'-ko')
ylabel('Cumulative Deaths')
xlabel('Date')
ylim([0,2000])
title('Identified Deaths in Massachusetts')
legend('Model','95% ci','95% ci','Observations','location','best')

```



When are the peaks?

Take the differences in the cumulative graphs and plot them up to see the peaks.

```

figure
yyaxis right
plot(Pred_dates(2:end),diff(Cov_Ma_Pred.deaths),'Color',[.94 .5 .5], 'LineStyle','-','Marker',
hold

```

Current plot held

```

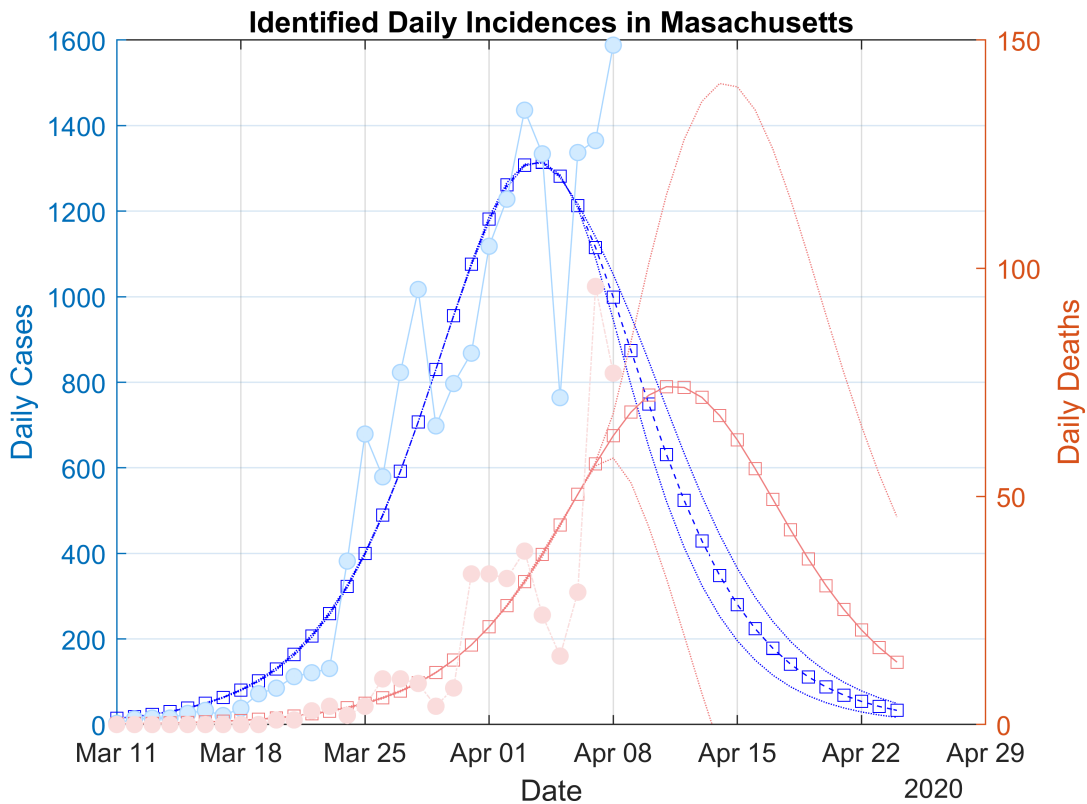
grid
plot(Pred_dates(2:end),diff(Cov_Ma_Pred.deathsci(:,1)),'Color',[.94 .5 .5], 'LineStyle',':')
plot(Pred_dates(2:end),diff(Cov_Ma_Pred.deathsci(:,2)),'Color',[.94 .5 .5], 'LineStyle',':')
plot(Cov_Ma.Date(2:end),diff(Cov_Ma.Deaths),'Color',[.98 .88 .88], 'MarkerFaceColor',[.98 .88
ylabel('Daily Deaths')
xlabel('Date')

```

```

ylim([0,150])
title('Identified Daily Incidences in Massachusetts')
%legend('Model','95% ci','95% ci','Observations', 'location', 'best')
yyaxis left
plot(Pred_dates(2:end),diff(Cov_Ma_Pred.cases),'--bs')
plot(Pred_dates(2:end),diff(Cov_Ma_Pred.casesci(:,1)),':b')
plot(Pred_dates(2:end),diff(Cov_Ma_Pred.casesci(:,2)),':b')
plot(Cov_Ma.Date(2:end),diff(Cov_Ma.Cases),'Color', [.68 .85 1], 'MarkerFaceColor', [.84 .92 1])
ylabel('Daily Cases')

```



```

%xlabel('Date')
%ylim([0,50])
%title('Identified Daily Incidences in Massachusetts')
%legend('Model','95% ci','95% ci','Observations', 'location', 'best')

```